

Date=17/07/2020

Lecture By=Shubham Joshi

Subject \Rightarrow Time Complexity

IN PREVIOUS LECTURE (QUICK RECAP) Date-14/07/2020	In Today's Lecture (Overview)
Revision Of Previous lectures For Python	<u>\Rightarrow SLA in Software</u> <u>\Rightarrow What Is Time Complexity In Python?</u> <u>\Rightarrow What Big O Notation Means?</u> <u>\Rightarrow What Are The Different Big O Notation Measures?</u> <u>$O(1)$:</u> <u>$O(\log n)$:</u> <u>$O(n)$:</u> <u>$(n \log n)$:</u> <u>$O(n \text{ square})$:</u> <u>\Rightarrow Examples</u>

⇒ SLA in Software

- A **software service level agreement (SLA)** is a contract between your business and your IT supplier.
- A service-level agreement (SLA) defines the level of service you expect from a vendor
- "[Click Here](#)" To know more about It

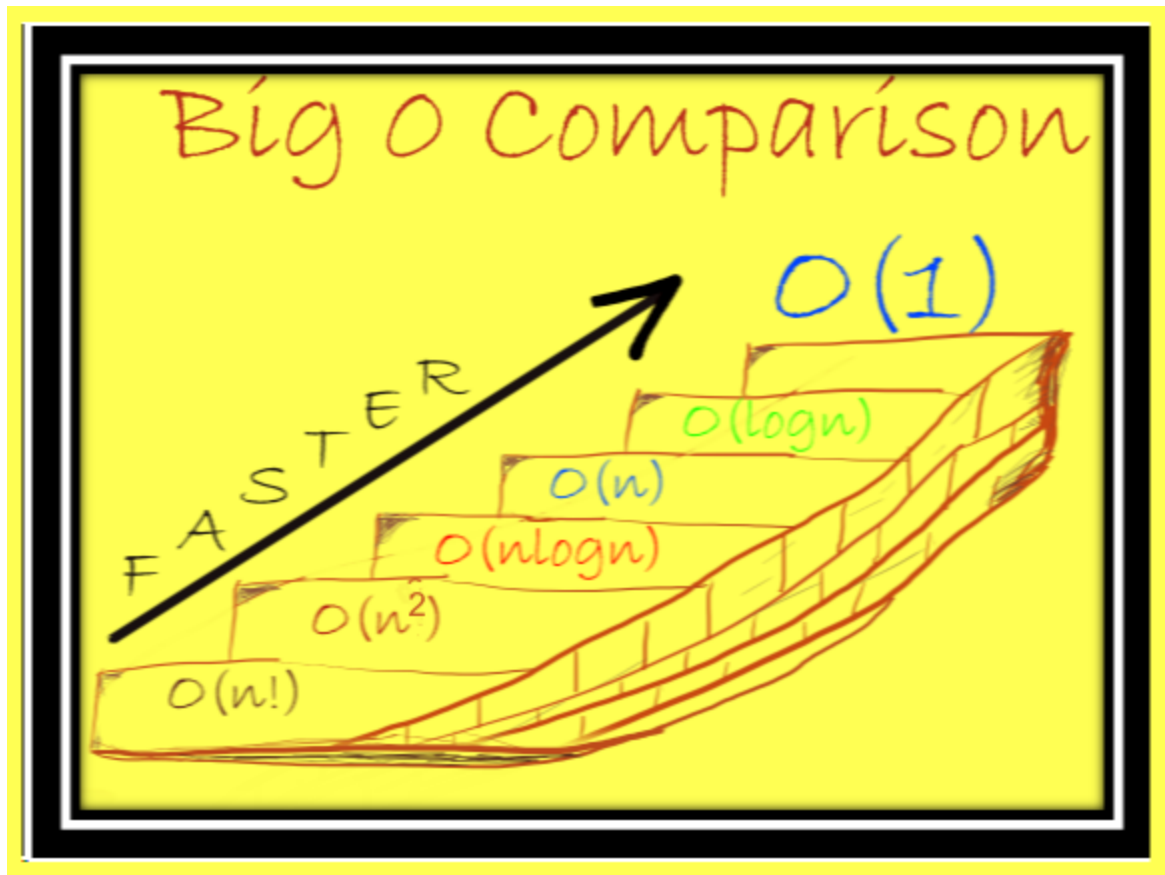
⇒ What Is Time Complexity In Python?

- Time complexity is commonly estimated by **counting the number of elementary operations performed by the algorithm**
- supposing that each elementary operation takes a fixed amount of time to perform.

⇒ What Big O Notation Means?

- A number of operations are performed in an algorithm.
- Big-O measures the time complexity of the operations of an algorithm.
- it measures how long it takes for the algorithm to compute the required operation.

==> **In simplest terms, Big O notation is a way to measure performance of an operation based on the input size, known as n.**



Big - O Notation	Computations for 10 Elements	Computations For 100 Elements	Computations For 1000 Elements
$O(1)$	1	1	1
$O(N)$	10	100	1000
$O(N^2)$	100	10000	1000000
$O(\log N)$	3	6	9
$O(N \log N)$	30	600	9000
$O(2^N)$	1024	1.26e+29	1.07e+301
$O(N!)$	3628800	9.33e+157	4.02e+2567

==>What Are The Different Big O Notation Measures?'

Name	Time Complexity
Constant Time	$O(1)$
Logarithmic Time	$O(\log n)$
Linear Time	$O(n)$
Quasilinear Time	$O(n \log n)$
Quadratic Time	$O(n^2)$

-**consider n to be the size of the input** collection. In terms of time complexity:

$O(1)$:

No matter how big your collection is, the time it takes to perform an operation is constant.

- As an instance, operations that check whether a collection has any items inside it is an $O(1)$ operation.

Example

```
if a > b:
    return True
else:
    return False
```

$O(\log n)$:

When the size of a collection increases, the time it takes to perform an operation increases logarithmically.

-This is the logarithmic time complexity notation. Potentially optimised searching algorithms are $O(\log n)$.

Example

```
for index in range(0, len(data), 3):  
    print(data[index])
```

$O(n)$:

The time it takes to perform an operation is directly and linearly proportional to the number of items in the collection.

-if we want to sum all of the items in a collection then we would have to iterate over the collection. Hence the iteration of a collection is an $O(n)$ operation.

Example

```
for value in data:  
    print(value)
```

$(n \log n)$:

Where the performance of performing an operation is a quasilinear function of the number of items in the collection.

- Time complexity of an optimised sorting algorithm is usually $n(\log n)$.

Example

```
for value in data1:  
    result.append(binary_search(data2, value))
```

O(n square):

When the time it takes to perform an operation is proportional to the square of the items in the collection. This is known as the quadratic time complexity notation

Example

```
for x in data:
    for y in data:
        print(x, y)
```

“[Click Here](#)” To Know More about it
“[Click Here](#)” for video Tutorial

⇒ Examples

```
1. a = 0 , i = N
   while i > 0 :
       a += i
       i /= 2
```

Complexity Of This Code is $\log n$

```
2. i = n / 2
   while i < n:
       j = 2
       while j < n:
           j *= 2
```

Complexity Of this Code is $N \log n$

```
3. count = 0 ;
   i = N
   while i > 0:
       While j < i:
           count += 1
           i /= 2 == Complexity Of this Code is O(N)
```